



Bayesian estimation of direct and correlated response to genetic selection on linear feed efficiency (RFI) or feed efficiency as a ratio trait (FCR)

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Efficiency in nutrient utilization by livestock is of great importance in animal breeding programs

Classical Measure of efficiency

output

input

Milk (Milk component)Yeild Dry matter intake

body weight gain feed consumption

In breeding programs: Inverse of efficiency

 $FCR = rac{Average\ daily\ feed\ intake}{Average\ daily\ gain}$







Disadvantages of ratio traits

LINEAR INDEX SELECTION TO IMPROVE TRAITS DEFINED AS RATIOS

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Difficulty in prediction of the effects of genetic selection on FCR in future generations as well as its effect on the production traits in the breeding program.

and (co)variances of its component traits (ADFI and ADG)





Linear feed efficiency

• Koch et al. (1963): RFI = FI – E(FI)

• E(FI) : based on BW and BWG and (maybe) carcass composition (LMP).

• Kennedy et al. (1993):

• IF genetic (co)variances are known for FI, BW, BWG, LMP. Genetic RFI can be estimated

$$\operatorname{var} \begin{bmatrix} y_r \\ y_p \end{bmatrix} = \begin{bmatrix} 1 & -b \\ 0 & 1 \end{bmatrix} \begin{bmatrix} \operatorname{var}(y_f) & \operatorname{cov}(y_f, y_p) \\ \operatorname{cov}(y_f, y_p) & \operatorname{var}(y_p) \end{bmatrix} \begin{bmatrix} 1 & 0 \\ -b & 1 \end{bmatrix}$$







Bayesian analysis

 Integrates unknown population parameters and "fixed" environmental effects and handles properly ratio traits

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Original article

Bayesian analysis of genetic change due to selection using Gibbs sampling

DA Sorensen¹, CS Wang², J Jensen¹, D Gianola²







Traits

- ADFI: 2.61 (0.39) kg/d
- ADG: 1.12 (0.15) kg/d
- LMP: 62.7 (1.97) percent



The MAXGRO[™] terminal line







Statistical model: Bayesian analysis

 $y_{ijkl}^{ADFI} = Gender_{ij} + Parity_{ik} + YS_{il} + b^{ADFI}(SBW_i) + a_i^{ADFI} + pe_i^{ADFI} + e_{ijkl}^{ADFI}$

 $y_{ijkl}^{ADG} = Gender_{ij} + Parity_{ik} + YS_{il} + b^{ADG}(SBW_i) + a_i^{ADG} + pe_i^{ADG} + e_{ijkl}^{ADG}$

 $y_{ijkl}^{LMP} = Gender_{ij} + Parity_{ik} + YS_{il} + b^{LMP}(EBW_i) + a_i^{LMP} + pe_i^{LMP} + e_{ijkl}^{LMP}$

Flat uninformative priors were assumed for all "fixed" effects in the model and for all (co)variance components.

$$\operatorname{Var}\begin{pmatrix} \mathbf{e}_{1} \\ \mathbf{e}_{2} \\ \mathbf{e}_{3} \end{pmatrix} = \mathbf{I} \otimes \mathbf{R}_{0} \qquad \qquad \operatorname{Var}\begin{pmatrix} \mathbf{a}_{1} \\ \mathbf{a}_{2} \\ \mathbf{a}_{3} \end{pmatrix} = \mathbf{A} \otimes \mathbf{G}_{0}$$







RFI estimation

Genetic RFI (RFIg)

$$a_i^{RFI_g} = a_i^{ADFI} - b_{g.1} a_i^{ADG} - b_{g.2} a_i^{LMP}$$

Phenotypic RFI (RFI_p)

$$a_i^{RFI_p} = a_i^{ADFI} - b_{p.1} a_i^{ADG} - b_{p.2} \hat{a}_i^{LMP}$$

$$\begin{bmatrix} \boldsymbol{G}_{RFI} & \boldsymbol{G}_{RFI,p} \\ \boldsymbol{G}_{p,RFI_g} & \boldsymbol{G}_p \end{bmatrix} = \boldsymbol{B} \boldsymbol{G} \boldsymbol{B}'$$
$$\begin{bmatrix} \boldsymbol{P}_{RFI} & \boldsymbol{P}_{RFI,p} \\ \boldsymbol{P}_{p,RFI} & \boldsymbol{P}_p \end{bmatrix} = \boldsymbol{B} \boldsymbol{P} \boldsymbol{B}'$$

$$B = \begin{bmatrix} \mathbf{I}_{fi} & -\mathbf{D}_{fip} \\ 0 & \mathbf{I}_p \end{bmatrix}; \mathbf{b}_{g_{fip}} = G^{-1}{}_p G_{p,fi}; \text{ or } \mathbf{b}_{p_{fip}} = P^{-1}{}_p P_{p,fi}$$

ECOFCE





Breeding values for FCR

The breeding value for FCR is dependent on the fixed effects since it depends on the mean of the traits involved in the ratio (μ_{ADFI} and μ_{ADG}).

$$a_i^{FCR} = \frac{\mu_{ADFI} + \hat{a}_i^{ADFI}}{\mu_{ADG} + \hat{a}_i^{ADG}} - \frac{\mu_{ADFI}}{\mu_{ADG}}$$

$$\mu_{ADFI} = \frac{\mathbf{1'b_1}}{rank(\mathbf{X})} + b_{1.1}\overline{S}$$
$$\mu_{ADG} = \frac{\mathbf{1'b_2}}{rank(\mathbf{X})} + b_{2.1}\overline{S}$$

where 1 is a vector of ones, rank(X) in this case is the number of elements in b_1 or b_2 ; and \overline{S} is the population average of start BW





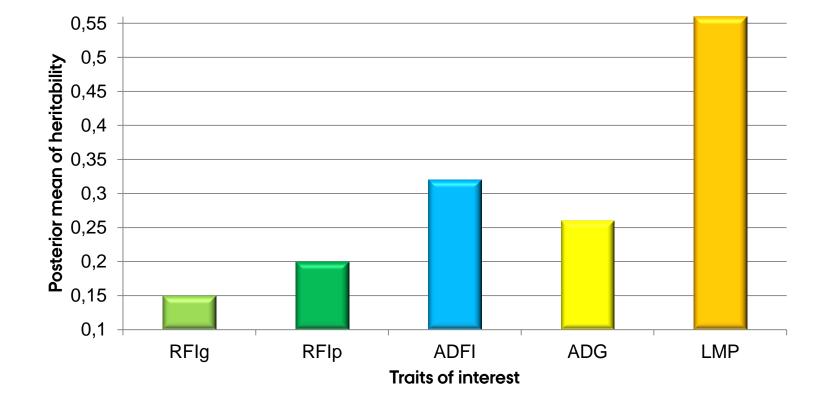


Bayesian response to selection

Difference between mean of EBV in the selected population compared to the whole population in each iteration = posterior distribution of response to selection







26.45% with 5.58 PSD of the genetic variation in ADFI is explained by RFI_g .

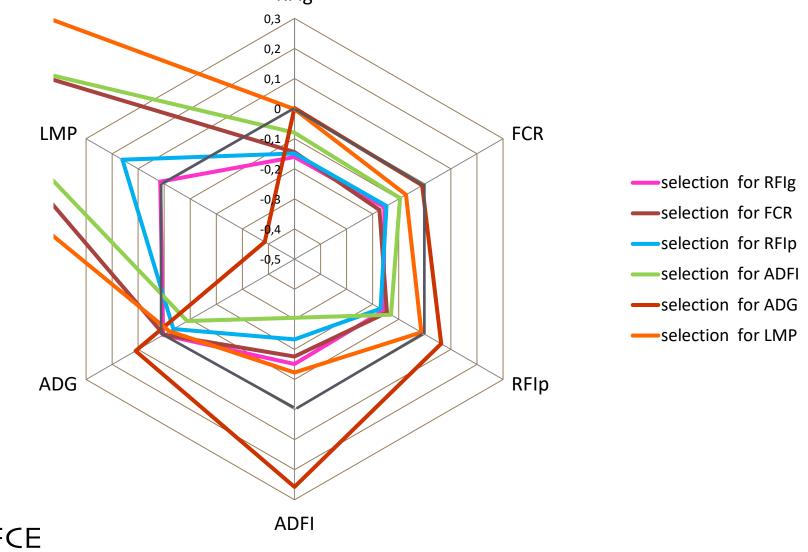


Selection for RFI results in reduction of ADFI without altering the production traits

Selection for FCR results in disproportional selection on its component traits RFIg

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Selection for FCR results in disproportional selection pressure on its component traits and also LMP

RFIg allows

- 1) selection on the proportion of feed intake that is independent of production; and
 - 2) easier and better selection index weights for the traits in the breeding programs

Selection for improved feed efficiency is likely best achieved through multiple-trait selection on RFI_g and production traits Joint selection for feed intake and production traits can also result in genetic improvement of feed efficiency





